

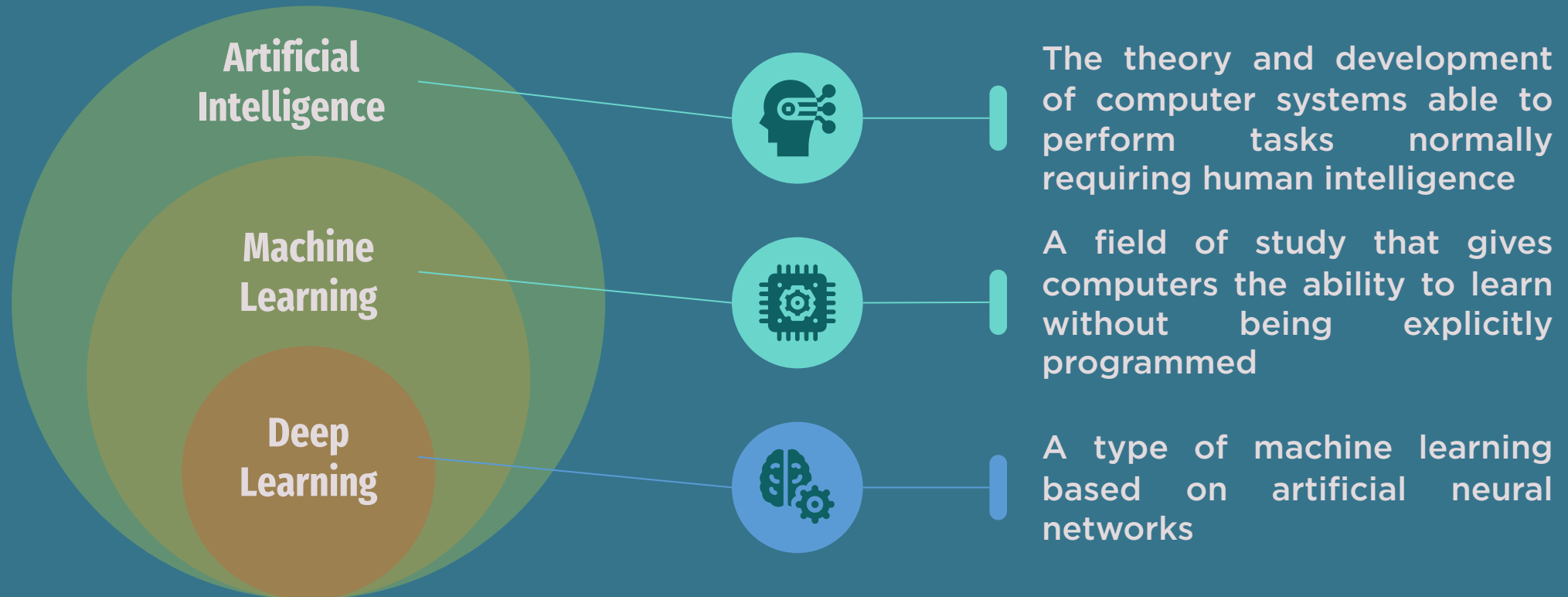


# Introduction to Machine Learning

**Bassem Ben Hamed**

*[bassem.benhamed@enetcom.usf.tn](mailto:bassem.benhamed@enetcom.usf.tn)*

# AI vs ML vs DL



# • Use cases for ML

## Healthcare

- Medical Imaging and diagnostics
- Personalized medicine
- Predictive approach to treatment

## Finance

- Algorithmic trading
- Fraud detection and prevention
- Portfolio management

## ML industry applications



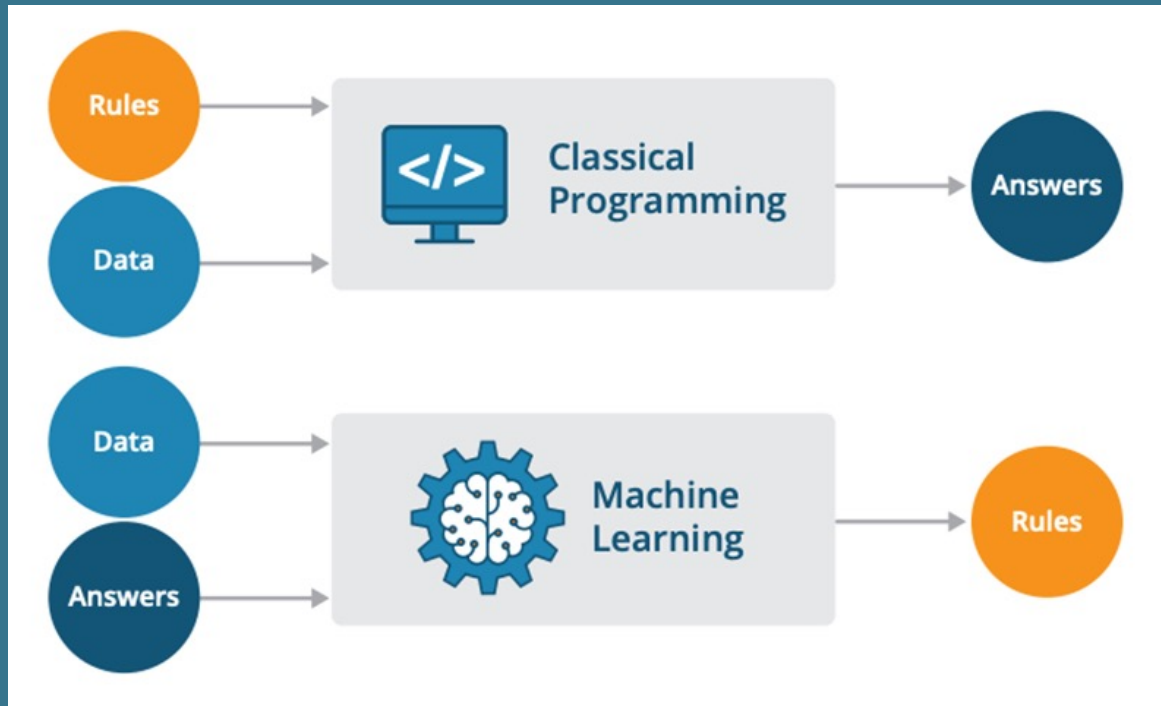
## Telecom

- Anomaly detection
- Predictive maintenance
- Churn prediction

## Manufacturing

- Energy consumption forecasting
- Predictive quality and yield
- Supply chain management

# Traditional Programming vs ML



Machine Learning: Field of study gives computers the ability to learn without being explicitly programmed.

Arthur Samuel

# • What is ML?

Learning is any process by which a system improves performance from experience.

Herbet Simon

Machine Learning is the study of algorithms that

- Improve their performance  $P$
- At some task  $T$
- With experience  $E$

As well-defined learning task is given by  $\langle P, T, E \rangle$ .

Tom Mitchell

# • Task vs Performance vs Experience

Improve on task T, with respect to performance metric P, based on experience E

T: Recognizing hand-written words

P: Percentage of words correctly classified

E: Database of human-labeled images of handwritten words

T: Categorize email messages as spam or legitime

P: Percentage of email messages correctly classified

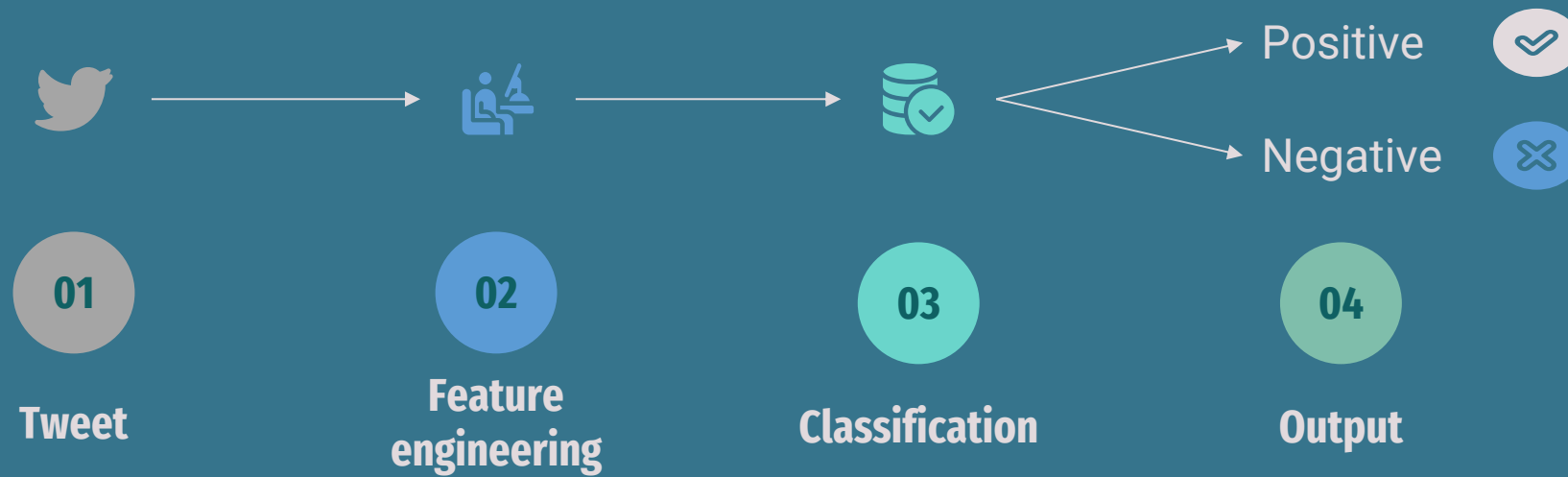
E: Database of emails, some with human-given labels

T: Playing checkers

P: Percentage of games won against an arbitrary opponent

E: Playing practice games against itself

# Twitter sentiment analysis example



# • Types of Learning

## Supervised (inductive) learning

Given training data + desired outputs (labels)

## Unsupervised learning

Given training data (without desired outputs)



## Semi-supervised learning

Given training data + a few of desired outputs

## Reinforcement learning

Rewards from sequence of actions



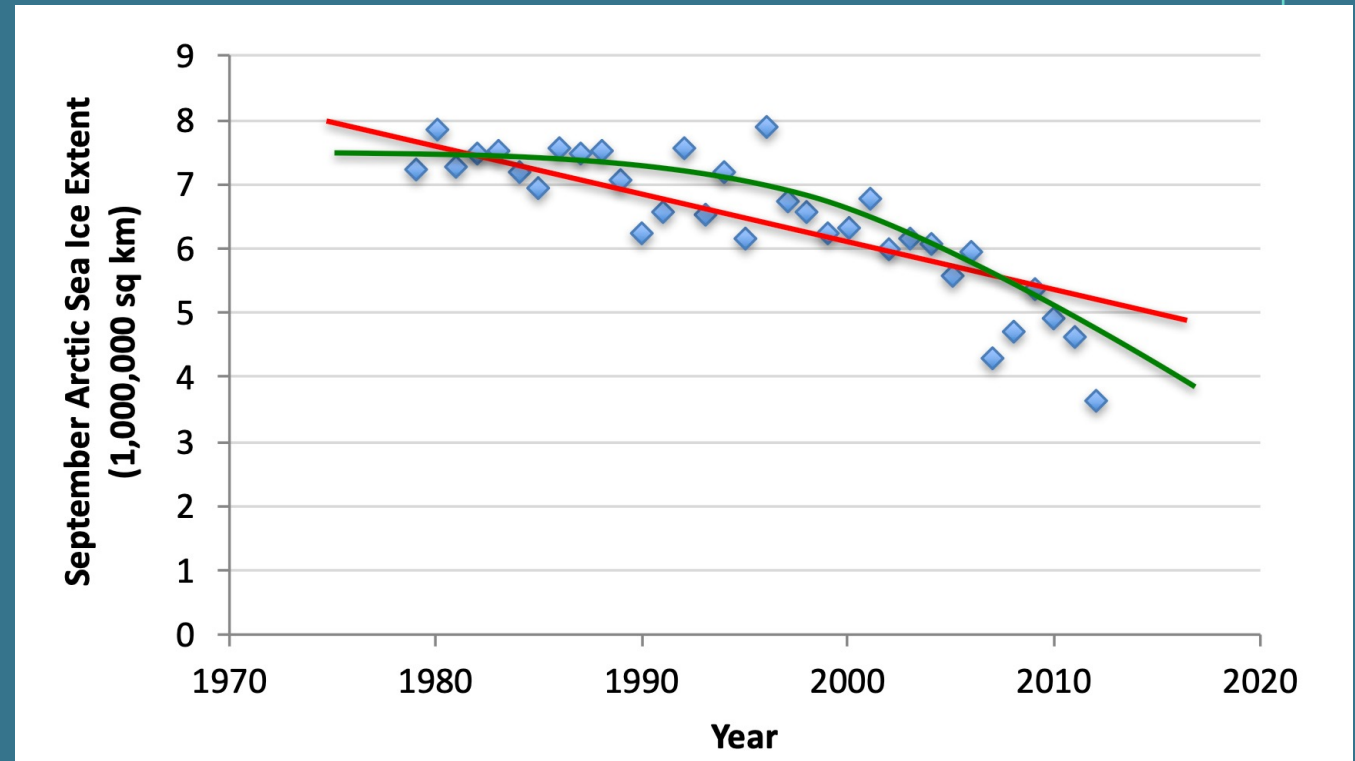
# Supervised Learning: Regression

Given

$(x_1, y_1), (x_2, y_2), \dots, (x_n, y_n)$

Learn a function  $f(x)$   
to predict  $y$  given  $x$

- $y$  is real-valued == regression



# Supervised Learning: Regression



Stock Price Prediction



Sales Forecasting



Housing prices



Risk Analysis

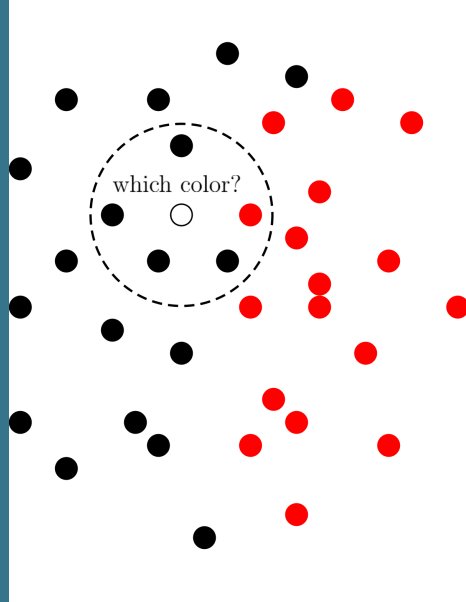
# Supervised Learning: Classification

Given  $(x_1, y_1), (x_2, y_2), \dots, (x_n, y_n)$

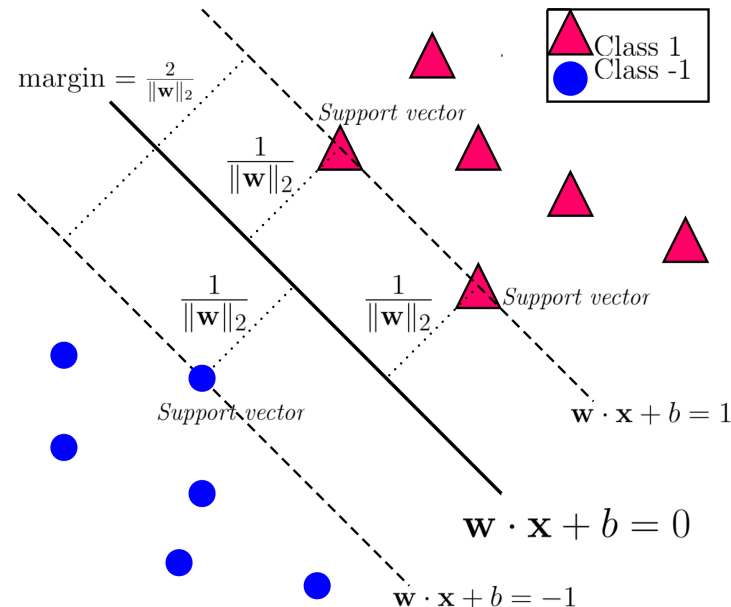
Learn a function  $f(x)$  to predict  $y$  given  $x$

- $y$  is categorical == classification

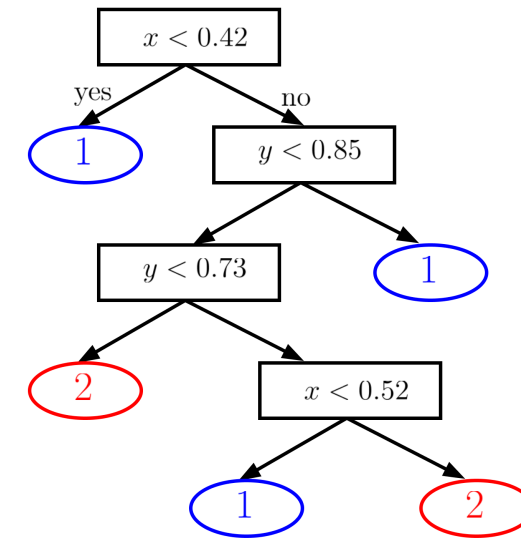
$k$ NN Classification



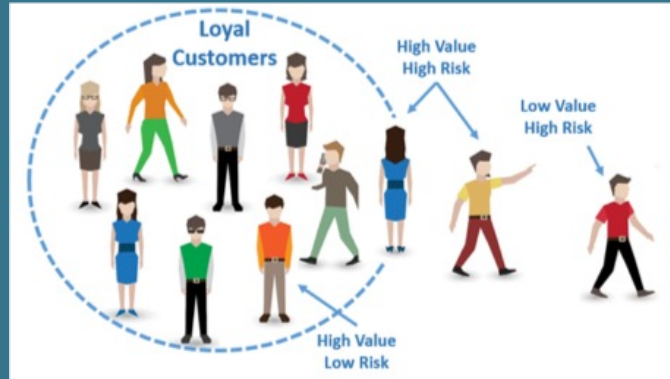
Support Vector Machine



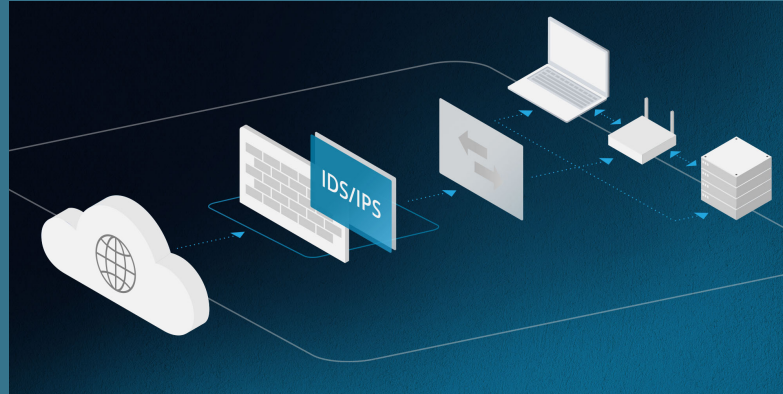
Classification Tree



# Supervised Learning: Classification



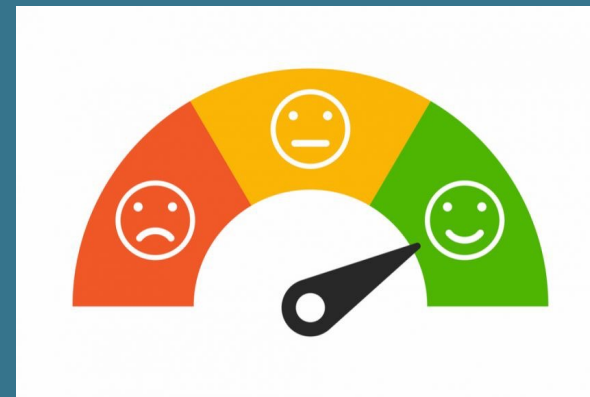
Customer Churn Prediction



Intrusion Detection System



Email Spam Detection



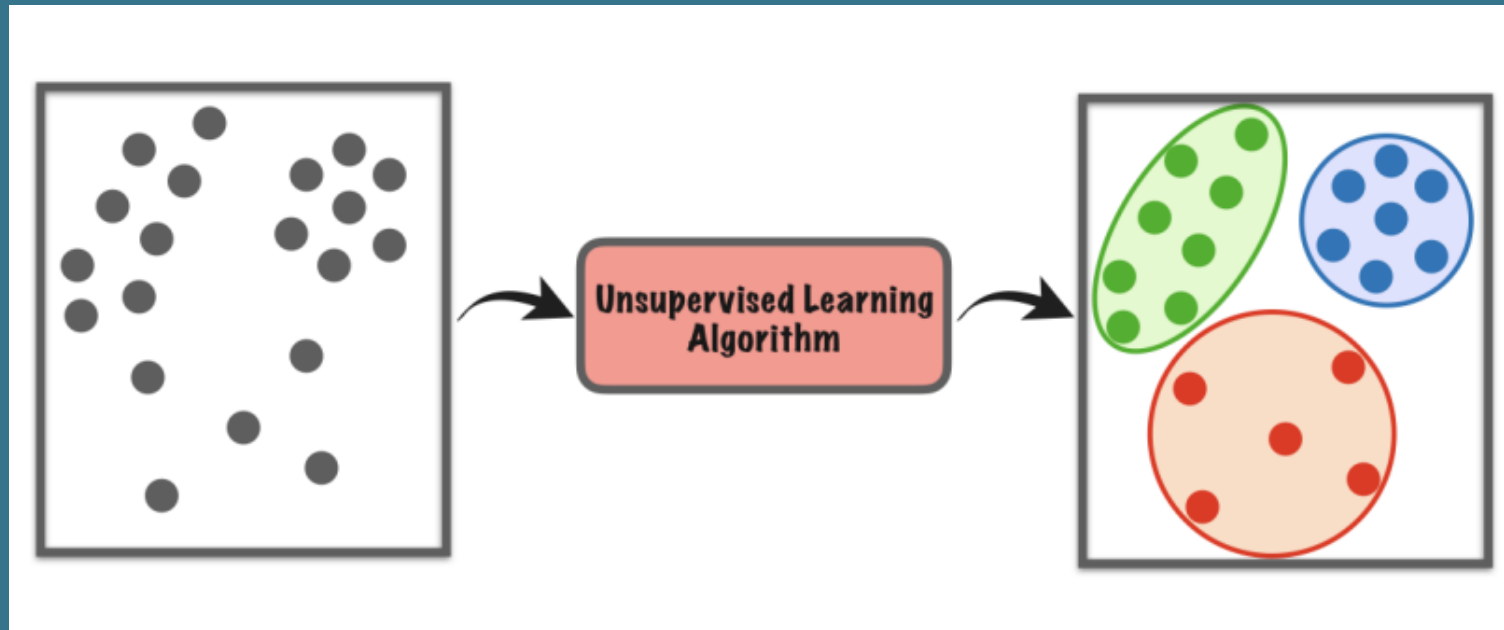
Sentiment Analysis

# • Unsupervised Learning

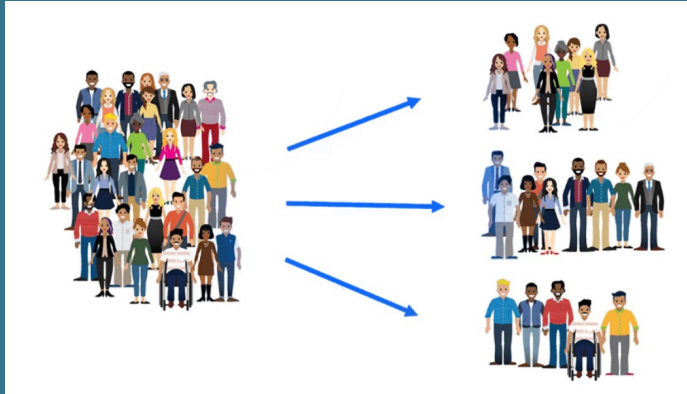
Given  $x_1, x_2, \dots, x_n$  (without labels)

Output hidden structure behind the  $x$ 's

- E.g., clustering



# Unsupervised Learning



Market segmentation



Organize computing clusters



Social network analysis



Group individuals by genetic similarity



# • Reinforcement Learning

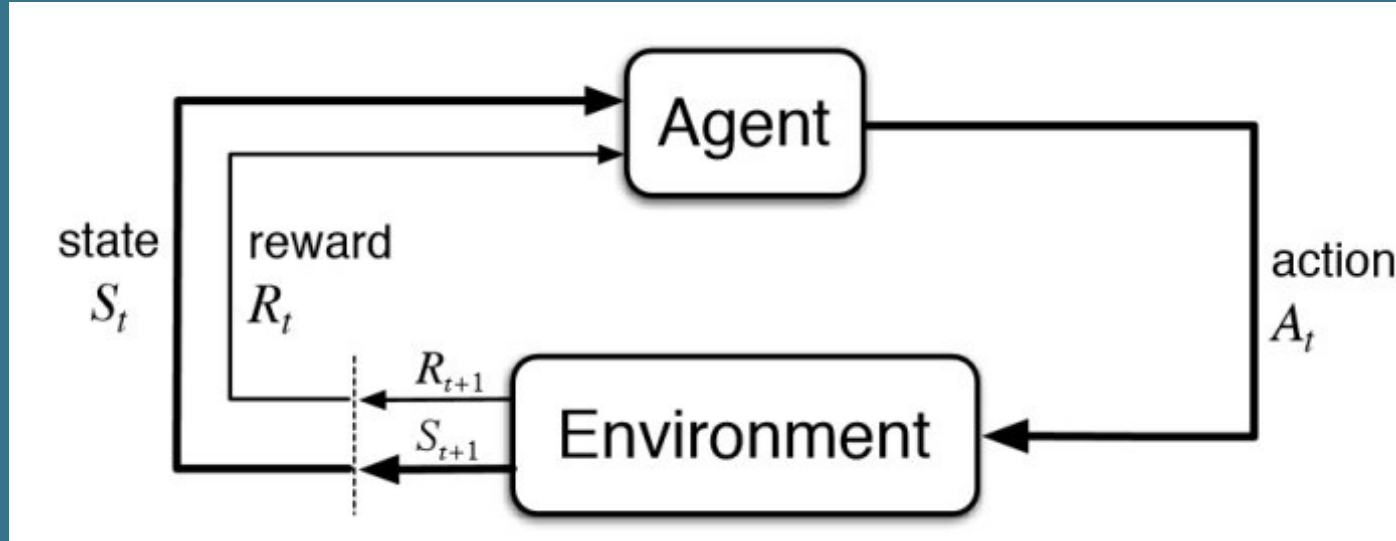
Given a sequence of states and actions with (delayed) rewards, output a policy

- Policy is a mapping from states to actions that tells you what to do in a given state

## Examples:

- Trading and finance
- News recommendation
- Natural Language Processing
- Healthcare
- Gaming
- Marketing and advertising

# Agent-Environment Interface

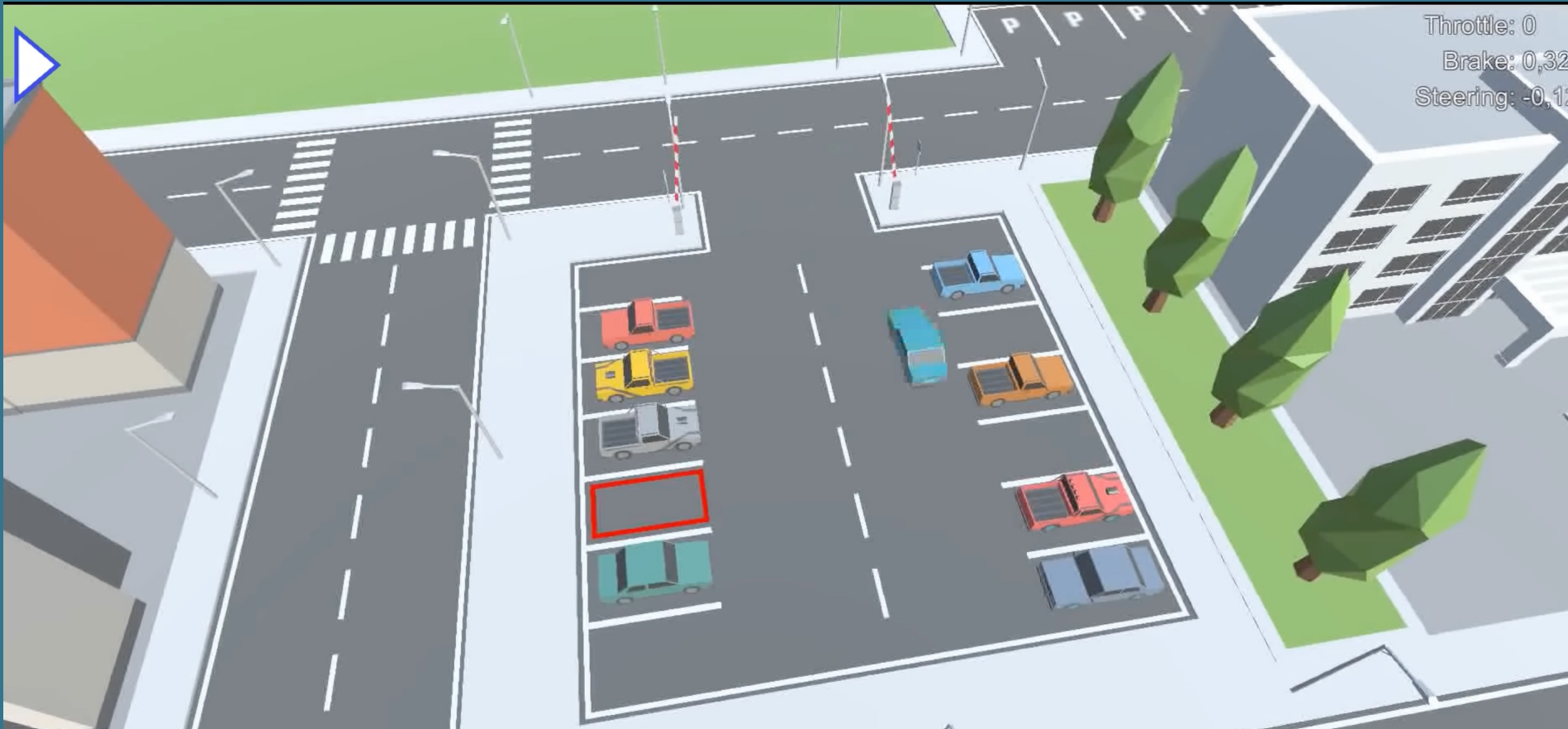


Agent and environment interact at discrete time steps:  $t = 0, 1, \dots, k$

Agent observes state at step  $t$ :  $s_t \in S$   
produces action at step  $t$ :  $a_t \in A(S)$   
gets resulting reward:  $r_{t+1} \in R$   
and resulting next state:  $s_{t+1}$



# AI learns to Park



[https://www.youtube.com/watch?v=VMp6pq6\\_QjI](https://www.youtube.com/watch?v=VMp6pq6_QjI)

# ML Life Cycle

## 1. Gathering Data

- Identify various data sources
  - Collect data
  - Integrate the data

## 2. Data Preparation

- Data exploration
- Data pre-processing

## 3. Data Wrangling

- Missing values
- Duplicate data
- Invalid data
- Noise

## 4. Data Analysis

- Selection of analytical techniques
- Building models
- Review the result

## 5. Train Model

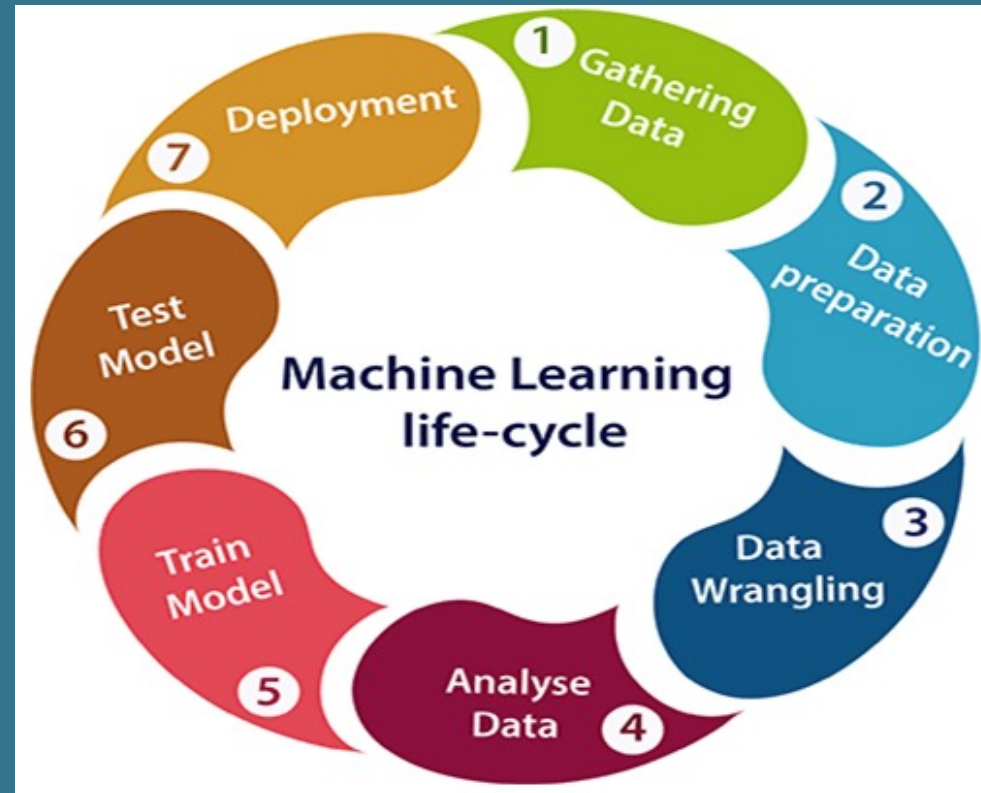
- Understand the various patterns, rules, and features

## 6. Test Model

- Testing the model and determines some metrics

## 6. Deployment

- Deploy the model in the real system



# References

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